

Identifying the Post COVID-19 Digital Divide in Chronic Pain Care

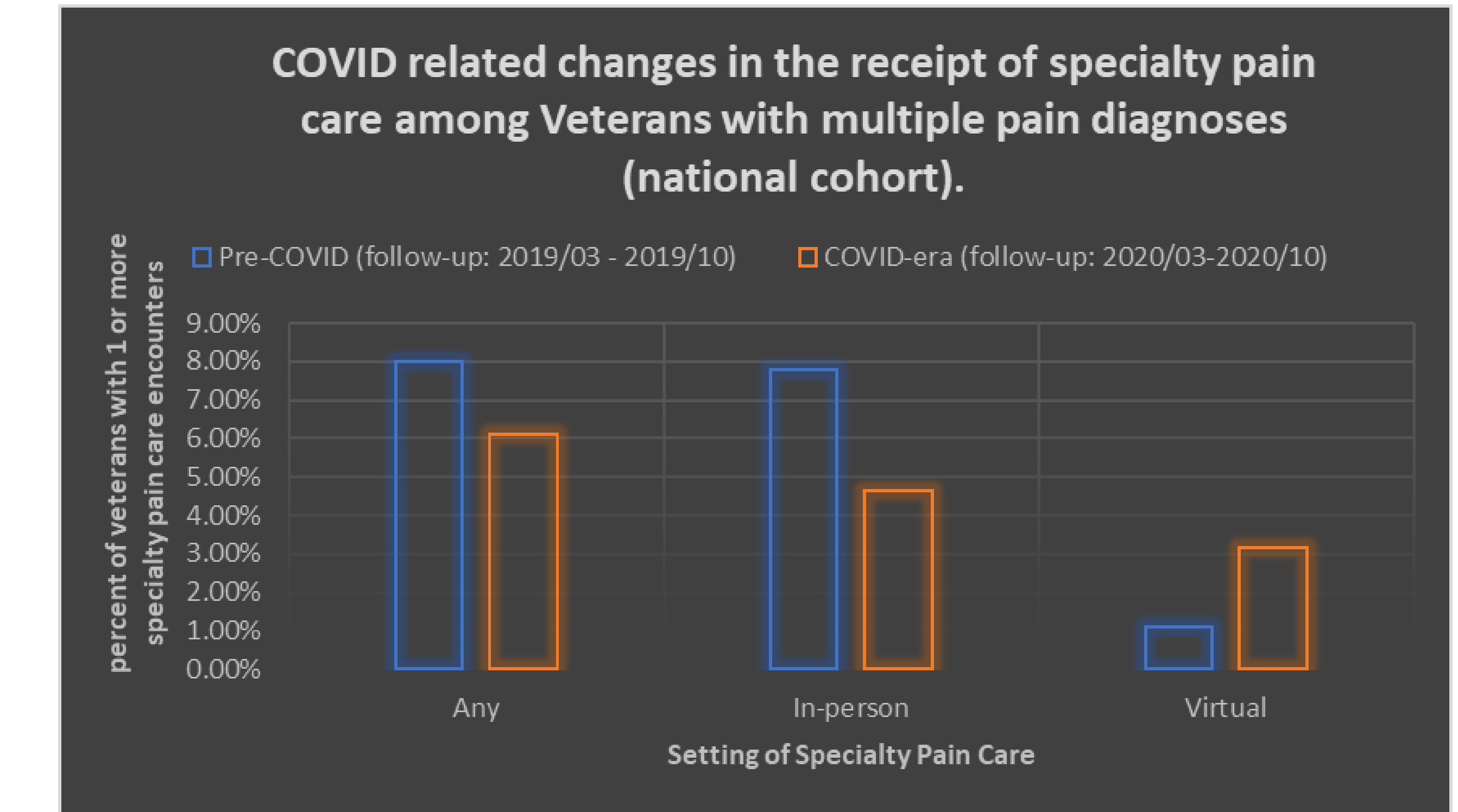


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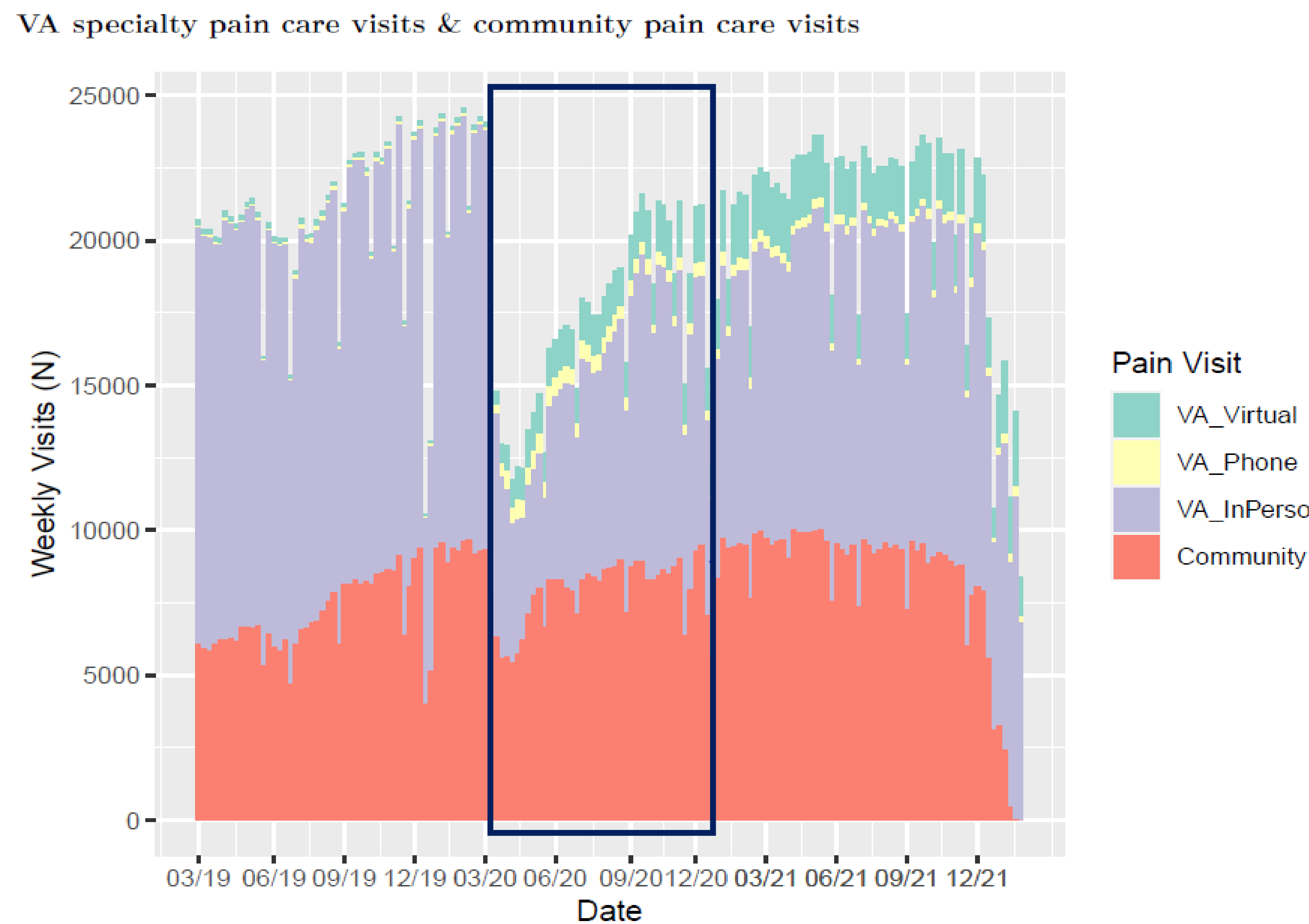
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WHAT HAPPENED TO SPECIALTY PAIN CARE UTILIZATION IN THE VA POST COVID ONSET ON AVERAGE?



- Between the pre-COVID and the COVID-era, the number of Veterans receiving specialty pain care:
 - Increased by 188.9% for virtual care
 - Decreased by 40.7% for in-person care
- During the 8 months post COVID onset, **24.4% fewer Veterans** received chronic pain specialty care in either virtual or in-person care.

PAIN SPECIALTY CARE IN THE COMMUNITY RESUMED FASTER THAN VA SPECIALTY PAIN CARE.

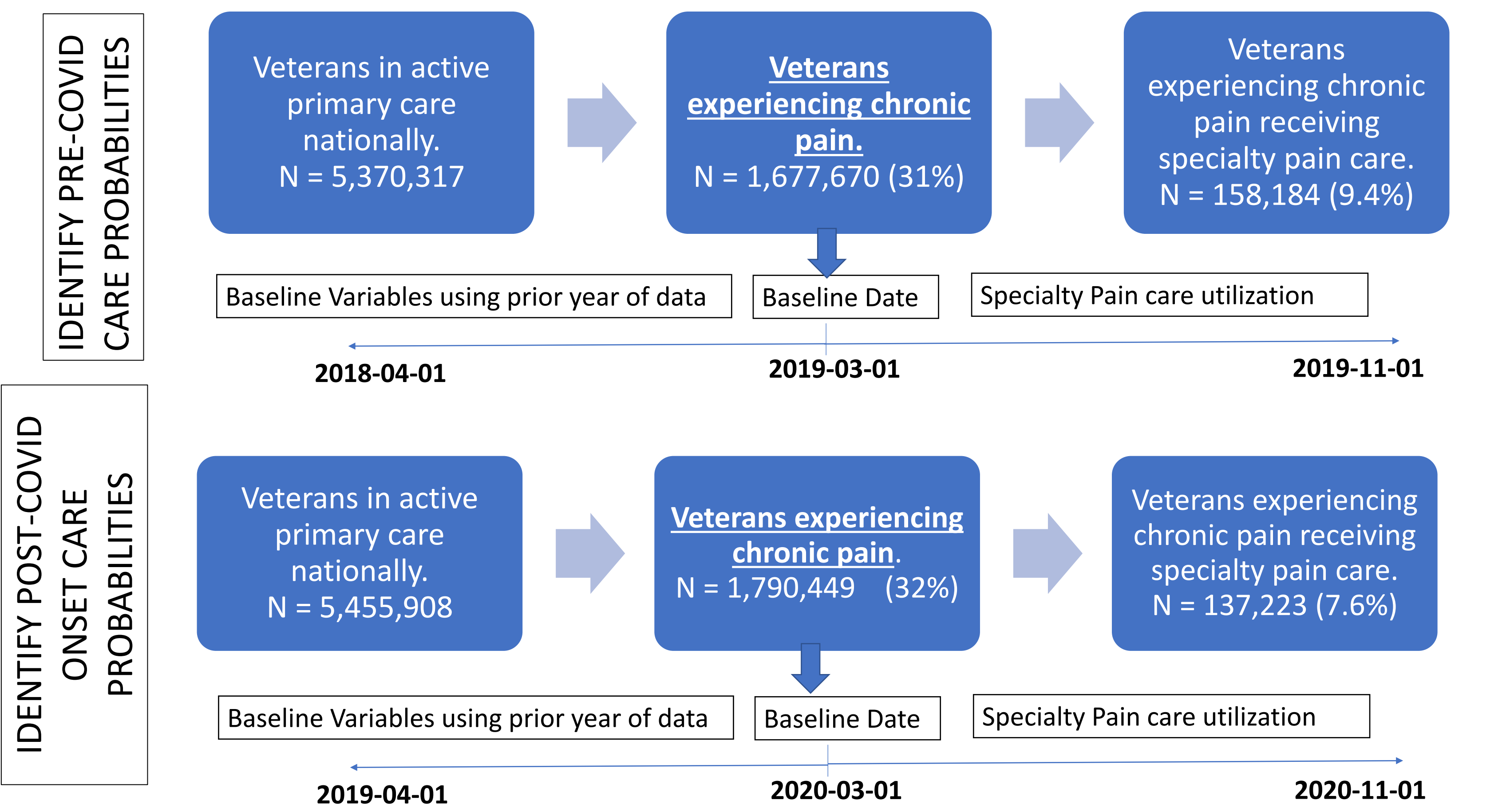


**Count of specialty pain visit days among all Veterans engaged in VA or VA paid community primary care.

OVERALL GOAL: BUILD AN ARTIFICIAL INTELLIGENCE (AI) SOLUTION TO IDENTIFY DEVELOPING CARE DISPARITIES.

- Identify Veterans at high risk of losing access to specialty pain care following the onset of COVID-19.
 - Construct national cohort of Veterans experiencing chronic pain.
 - Build relevant computational phenotypes.
 - Use machine learning approaches to identify expected probability of pain specialty care utilization using pre-COVID and post COVID onset cohorts.
- Identify Veterans at highest risk of losing access to care.
- Compare Denver, Seattle, and National results.

STUDY DESIGN



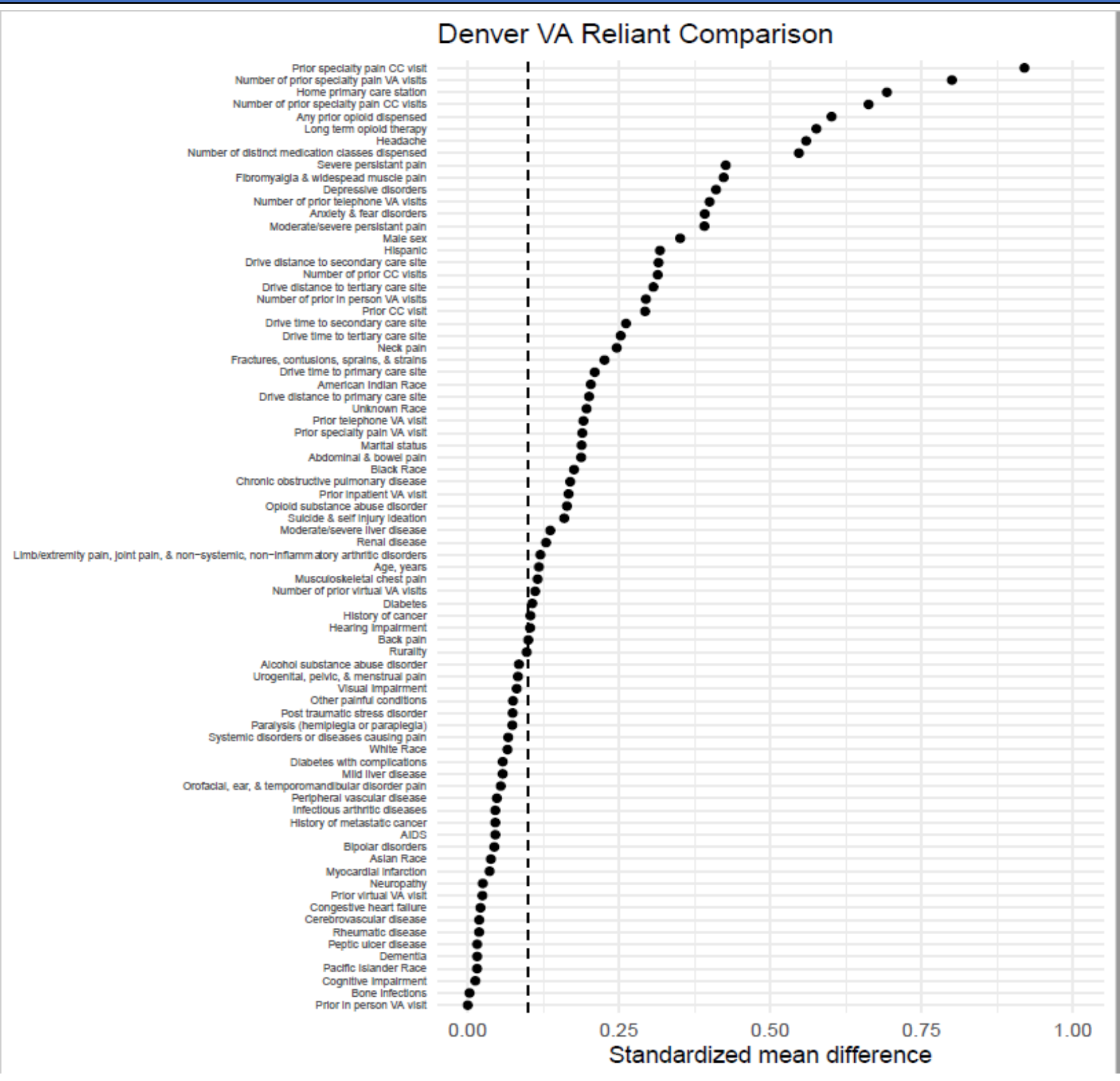
MACHINE LEARNING MODELING APPROACH¹

- Goal:** Conditional on relevant baseline variables, estimate expected probability of care utilization if COVID had never occurred.
 - ML Model 1:** Using the *pre-COVID cohort* to represent this counterfactual state, develop ML model estimating probability of care engagement.
 - After developing ML model 1, make predictions using the post COVID onset cohort.
- Goal:** Conditional on relevant baseline variables, estimate expected probability of care utilization during COVID.
 - ML Model 2:** Using the *post-COVID onset cohort* to represent this observed state, develop ML model estimating probability of care engagement post-COVID.
- The difference in predicted probability between ML Model 1 and ML Model 2 is the estimated change in access to care.
- Models considered:** GBM, XGBoost, GLM, Deep Learning, DRF, stacked ensemble all models (42 total models tried representing different hyperparameter settings)
- Model Validation:** 5-fold cross validation with external assignment into folds prior to model fitting.
- F1 statistic** (harmonic average of Precision and Recall) chosen for model optimization/selection due to somewhat rare outcome (true negatives not of interest).
- Multiple models fit:**
 - For each of (National), (Seattle), (Denver):
 - Pre-covid model using pre-covid cohort
 - Post-covid onset model using post covid onset cohort

Focus on Denver Area Veterans: VA Reliant group.

23,534 Denver Veterans experiencing chronic pain during COVID era. **1,132** had a predicted probability of pre-COVID VA pain care of at least 0.2.

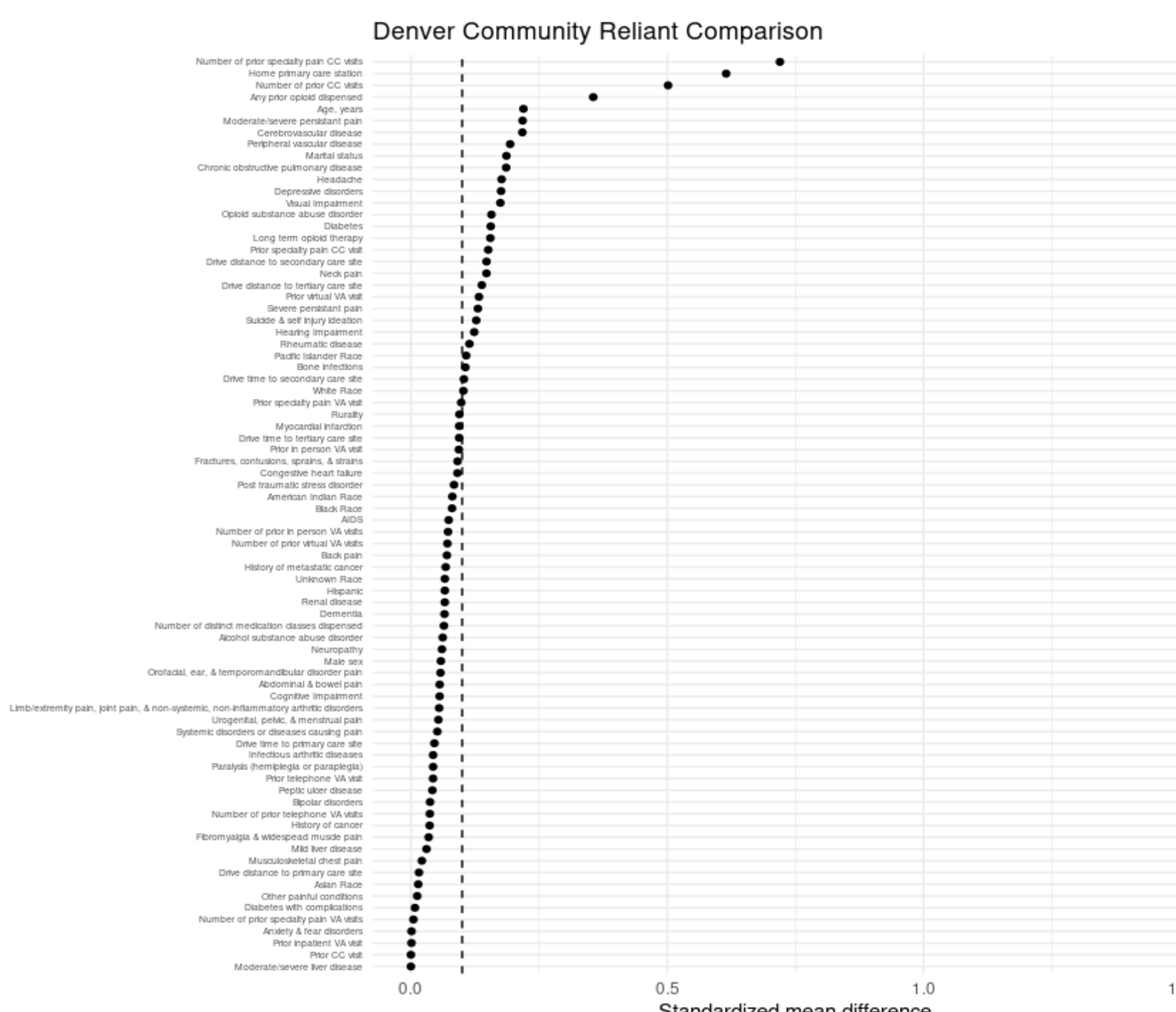
	Prob diff <=0	0< Prob diff <=0.25	Prob diff > 0.25
n	103	745	284



Focus on Denver Area Veterans: Community Reliant group.

23,534 Denver Veterans experiencing chronic pain during COVID era. **1,586** had a predicted probability of pre-COVID VA pain care of at least 0.2.

	Prob diff <=0	0< Prob diff <=0.25	Prob diff > 0.25
n	449	1,067	70



Results Summary

VA reliant:	Community reliant:
<ul style="list-style-type: none"><u>Less likely to lose access:</u><ul style="list-style-type: none">Veterans exposed to Opioids.White or FemaleYounger VeteransVeterans with MH comorbiditiesVeterans with Pain related diagnosesVeterans with SUD<u>More likely to lose access:</u><ul style="list-style-type: none">Black or Hispanic VeteransOlder VeteransUrban Veterans	<ul style="list-style-type: none">Gender/race not associated with access<u>Less likely to lose access:</u><ul style="list-style-type: none">Veterans exposed to Opioids.Veterans with MH comorbiditiesVeterans with Pain related diagnoses<u>More likely to lose access:</u><ul style="list-style-type: none">Veterans with MH comorbiditiesVeterans with SUD

ACKNOWLEDGEMENTS

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